INFO 6210 Data Management and Database Design Al Skunkworks Project | Hyperparameter Database | DB14

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Abstract

Gather a list of data sets, type of datasets, and hyperparameters by running an expanded list of datasets. This information will be embedded in a database management system, to be incorporated into a website where it is easy to be searched and used by the public.

The hyperparameter database is created by running millions of hyperparameter values, over thousands of public datasets and calculating the individual conditional expectation of every hyperparameter on the quality of a model.

Generate models using H2O software to find the best hyperparameters and create a conceptual model and store all the data into a physical database.

Introduction

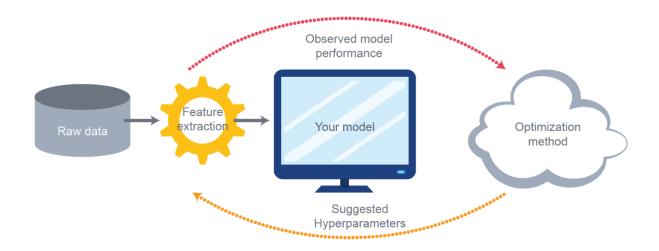
Hyperparameters are parameters that are specified prior to running machine learning algorithms that have a large effect on the predictive power of statistical models. Knowledge of the relative importance of a hyperparameter to an algorithm and its range of values is crucial to hyperparameter tuning and creating effective models.

The hyperparameter database is a public resource with algorithms, tools, and data that allows users to visualize and understand how to choose hyperparameters that maximize the predictive power of their models.

The hyperparameter database is created by running millions of hyperparameter values, over thousands of public datasets and calculating the individual conditional expectation of every hyperparameter on the quality of a model.

Currently, the hyperparameter database analyzes the effect of hyperparameters on the following algorithms: Distributed Random Forest (DRF), Generalized Linear Model (GLM), Gradient Boosting Machine (GBM). Naïve Bayes Classifier, Stacked Ensembles, Xgboost and Deep Learning Models (Neural Networks).

The hyperparameter database also uses these data to build models that can predict hyperparameters without search and for visualizing and teaching statistical concepts such as power and bias/variance tradeoff.



Objective

To create Hyperparameter Database by running several hyperparameter values on several datasets and to calculate the individual conditional expectations on every hyperparameter on quality of model



Project Requirement:

Unique datasets are to be picked from different data sources like Kaggle Datasets, UCI machine learning repository, Amazon Datasets, Google Datasets, Computer Vision Datasets etc. Identify the type of dataset chosen, ie Regression, Classification, Clustering etc. Perform data cleaning

and data pre-processing. Create conceptual and ER diagrams. Perform database normalization and perform analytics on the database created to get the best values for the hyperparameters.

Problems to be addressed:

Most of the algorithms that improve metrics, degrades the quality of search results. Hyperparameter optimization is performed to overcome the issues addressed by those algorithms and build models for visualizing and teaching statistical concepts.

Potential pitfalls & challenge:

Different optimization methods will have different setup steps, time requirements, and performance outcomes. Hence, methods like algorithmic optimization will help in achieving better performance.

Data source

Data is derived from Kaggle Datasets. We have chosen considered dataset of Travel Insurance.

https://www.kaggle.com/mhdzahier/travel-insurance

A third-party TRAVEL INSURANCE servicing company that is based in Singapore.

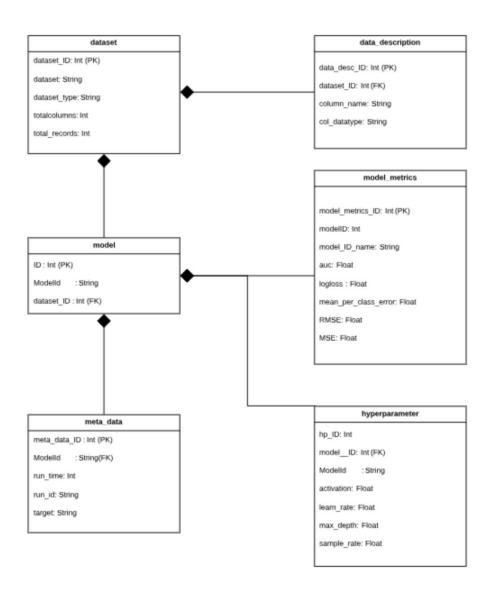
The attributes:

- Target: Claim Status
- Name of agency
- Type of travel insurance agencies
- Distribution channel of travel insurance agencies
- Name of the travel insurance products
- Duration of travel
- Destination of travel
- Amount of sales of travel insurance policies
- Commission received for travel insurance agency
- · Gender of insured
- Age of insured

Conceptual Schema

A conceptual data model identifies the highest-level relationships between the different entities.

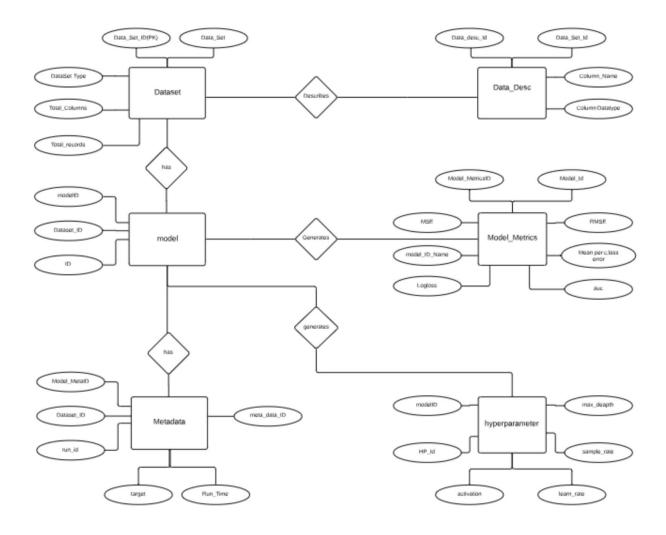
- dataset_ID is the primary key of dataset table.
- data_desc_ID is the primary key and dataset_ID is the foreign key of data_description table.
- meta data ID is the primary key and dataset ID is the foreign key of meta data table.
- model_metrics_ID is the primary key and meta_data_ID is the foreign key of model_metrics table.
- hp_ID is the primary key and model_metrics_ID is the foreign key of hyperparameter table.



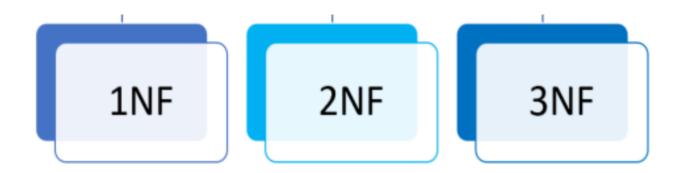
ER Diagram

An entity relationship diagram shows the relationships of entity sets stored in a database. An entity is an object, a component of data. An entity set is a collection of similar entities. These entities have attributes that define its properties.

Entity dataset describes the entity data_description and has the entity meta_data. Entity meta_data generates entity model_metrics. Entity model_metrics generates hyperparameter entity.



Normalization



According to 1 Normal Form (1NF),

- 1. There are no repeating groups
- 2. Maintained atomic data values of hyperparameter table is further split into hyperparameter default and hyperparameter actual columns
- 3. Each field of the table has unique name
- 4. Each table has primary key

According to 2 Normal Form (2NF),

- 1. All tables satisfy 1NF
- 2. All non-key attributes are dependent on all parts of primary key. Thus, no partial dependencies
- 3. There are no calculated data
- 4. Each field of the table has unique name
- 5. Each table has primary key

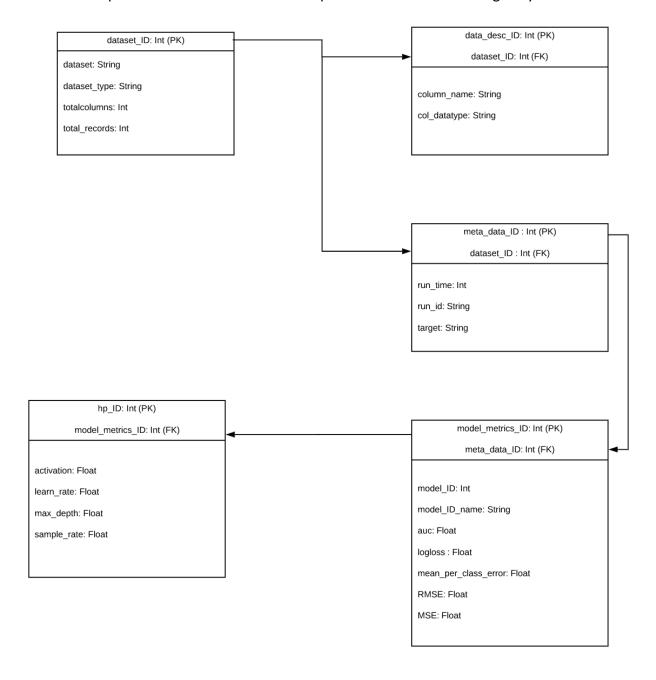
According to 3 Normal Form (3NF),

- 1. All tables satisfy 2NF
- 2. All non-key attributes are not dependent on other non-key attributes. Thus, no transitive relationship
- 3. Each field of the table has unique name
- 4. Each table has primary key

Physical Model

Physical data model represents how the model will be built in the database. A physical database model shows all table structures, including column name, column data type, column constraints, primary key, foreign key, and relationships between tables.

All the entities from ER diagram is converted to tables. All the attributes are converted into columns of respective tables. All the relationships are converted into foreign keys.



Use Cases

- 1. Number of 'run time' for dataset 'Travel Insurance'
- 2. Number of models for run time '1000'
- 3. Return the actual value of learn_rate for model ID 1
- 4. What are the metric values of the learn rate of model 1 of run time 1000
- 5. Which run time has the highest RMSE value
- 6. Which run time has the least logloss value
- 7. Total number of models generated for all the runs
- 8. Auc value for model 3 of run time 1500
- 9. Display the number of models for each run times
- 10. Which Run has the best model
- 11. For run time 1500 which model ID has the ideal AUC value

Analytics

Below are the queries for the above use cases





Functions

A function is a special kind stored program that returns a single value. Functions are used to encapsulate common formulas or business rules that are reusable among SQL statements or stored programs. It helps improve the readability and maintainability of the procedural code.

The following illustrates the syntax for creating a new function:

CREATE FUNCTION function_name(param1,param2,...)
RETURNS datatype
[NOT] DETERMINISTIC
Statements

Below are the functions for the database:

```
Return run_id for highest run time
 42
 43
         DELIMITER $$
 44 •
         CREATE FUNCTION FUN1()
 45
         RETURNS TEXT
         DETERMINISTIC

→ BEGIN

         DECLARE func TEXT;
 48
 49
         select run id INTO func from hyperparameter db14.meta data where run time=
 50
         (select max(run_time) from hyperparameter_db14.meta_data);
         RETURN func;
 51
         END $$
 52
 53
         DELIMITER;
        select FUN1();
Export: Wrap Cell Content: $\overline{A}$
   FUN1()
3oTEV9hn4
```

```
60
         -- 4. Return dataset id for highest run time
  61
        DELIMITER $$
  62 • CREATE FUNCTION FUN2()
  63
        RETURNS INTEGER
  64
       DETERMINISTIC
  65 ⊝ BEGIN
  66
       DECLARE func integer;
  67 select dataset_id into func from hyperparameter_db14.meta_data where run_time= (select max(run_time)
  68
        from hyperparameter_db14.meta_data);
        RETURN func;
  69
       END $$
  70
        DELIMITER ;
  71
  72 • select FUN2();
<
Export: Wrap Cell Content: TA
  FUN2()
) 1
        -- 5. Highest logloss value for model id 5 for run time 2500
 79
        DELIMITER $$
 81 • CREATE FUNCTION FUN3()
        RETURNS float
 82
        DETERMINISTIC
 83
 84 ⊖ BEGIN
       DECLARE func float;
 85
        select max(logloss) INTO func from hyperparameter_db14.model_metrics mm, hyperparameter_db14.meta_data md
 86
        where mM.meta data ID=md.meta data ID and model ID=5 and run time=2500;
 87
        RETURN func;
 88
      END $$
 89
 90
        DELIMITER;
 91 • select FUN3();
<
Export: Wrap Cell Content: IA
   FUN3()
0.06681600958108902
 100
        -- 6. Number of models for run time 2000
        DELIMITER $$
 101
 102 • CREATE FUNCTION FUN4()
 103
        RETURNS INTEGER
        DETERMINISTIC
 104
 105 ⊝ BEGIN
 106
        DECLARE func integer;
        select count(model_ID) INTO func from hyperparameter_db14.model_metrics mm, hyperparameter_db14.meta_data md
 107
 108
        where mm.meta_data_ID=md.meta_data_ID and run_time=2000;
        RETURN func;
 109
       END $$
 110
        DELIMITER;
 112 •
        select FUN4();
<
                                    Export: Wrap Cell Content: IA
Result Grid 🔠 🙌 Filter Rows:
                                                                                                         FUN4()

• 47
```

Procedures

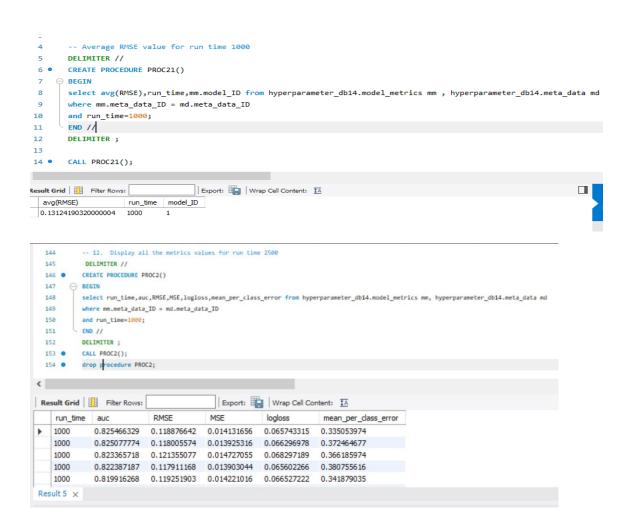
A stored procedure is a segment of declarative SQL statements stored inside the database catalog. A stored procedure can be invoked by triggers, other stored procedures, and applications such as Java, Python, PHP. Stored procedures increase the performance of the applications.

The following illustrates the syntax for creating a new function:

CREATE

```
[DEFINER = user]
PROCEDURE sp_name ([proc_parameter[,...]])
[characteristic ...] routine body
```

Below are the procedures for the database:



```
27
           -- 13. What are the start and end time of all runs
   28
           DELIMITER //
           CREATE PROCEDURE PROC3()
   29 •
   30

⊕ BEGIN

           select start_time, end_time, run_time from hyperparameter_db14.meta_data;
   31
         END //
   32
   33
           DELIMITER;
           CALL PROC3();
   34 •
 <
                                          Export: Wrap Cell Content: IA
 Result Grid Filter Rows:
     start_time
                 end_time
                             run_time
     1555918564 1555918564
                             500
     1555920195 1555920195 1000
     1555968866
                1555968866
                             1500
     1555972904 1555972904 2000
    1555975812 1555975812 2500
  18
        -- Display MSE value for model 10 of run time 1000
  19
       DELIMITER //
  20 • CREATE PROCEDURE PROC24()
  select MSE, mm.model_ID, run_time from hyperparameter_db14.model_metrics mm , hyperparameter_db14.meta_data md
  22
  23
       where mm.meta_data_ID = md.meta_data_ID
  24
       and run_time=1000;
      END //
 25
      DELIMITER;
 27 • CALL PROC24();
<
Result Grid | II Filter Rows:
                              Export: Wrap Cell Content: IA
                                                                                                 MSE
            model_ID run_time
0.014131656
                    1000
                 1000
   0.013925316 2
   0.014727055
                   1000
 20
 29
         15. Highest MSE value for run time 500
         DELIMITER //
 31 •
         CREATE PROCEDURE PROC25()

→ BEGIN

 32
         select max(MSE), run_time,mm.model_ID from hyperparameter_db14.model_metrics mm ,
 33
 34
         hyperparameter db14.meta data md
 35
         where mm.meta_data_ID = md.meta_data_ID and run_time=1000;
         END //
 36
         DELIMITER;
 37
         CALL PROC25();
 38 •
                                        Export: Wrap Cell Content: IA
Result Grid Filter Rows:
   max(MSE)
               run_time | model_ID
▶ 0.05530114
               1000
```

What are the run id of model id that has least mean_per_class_error

```
DELIMITER //
  64 •
         CREATE PROCEDURE PROC6()
      ⊖ BEGIN
   65
          select run_id from hyperparameter_db14.meta_data_md, hyperparameter_db14.model_metrics mm
   66
       where mm.meta_data_ID = md.meta_data_ID AND mean_per_class_error= (select min(mean_per_class_error))
   67
   68
          from hyperparameter_db14.model_metrics);
         END //
         DELIMITER ;
         CALL DDOCK().
 Result Grid Filter Rows:
                                   Export: Wrap Cell Content: A
    run id
 ▶ PVxu8FKke
   دد
               -- Return run_id for lowest run time
           DELIMITER //
   55
          CREATE PROCEDURE PROC27()

⊖ BEGIN

   58
           select run_id, run_time lowest_run_time from hyperparameter_db14.model_metrics mm ,
           hyperparameter_db14.meta_data md
   59
           where mm.meta_data_ID = md.meta_data_ID
   60
           and run_time= (select min(run_time) from hyperparameter_db14.meta_data);
   62
           END //
           DELIMITER;
   63
           CALL PROC27();
   64 •
 Result Grid Filter Rows:
                                        Export: Wrap Cell Content: IA
     run_id
                lowest_run_time
 wtiOmXHNa
 66
              Return dataset_id for lowest run time
 67
         DELIMITER //
         CREATE PROCEDURE PROC28()
 68 •

⊖ BEGIN

 69
         select dataset_id, run_time lowest_run_time from hyperparameter_db14.meta_data
 70
 71
          where run_time= (select min(run_time) from hyperparameter_db14.meta_data);
 72
       END //
         DELIMITER;
 73
       CALL PROC28();
 74 •
                                       Export: Wrap Cell Content: IA
Result Grid | Filter Rows:
   dataset_id lowest_run_time
▶ 1
             500
```

```
100
          -- 19. Display the parameters for travel insurance dataset.
 101
          DELIMITER //
          CREATE PROCEDURE PROC9()
 102 •
 103
          select dataset ID, dataset dataset Name, dataset type, total columns, total records from dataset;
 104
          END //
 105
          DELIMITER;
 106
          CALL PROC9();
 107 •
 Result Grid Filter Rows:
                                        Export: Wrap Cell Content: TA
    dataset_ID dataset_Name
                                 dataset_type
                                             total_columns
                                                          total_records
) 1
              TRAVEL INSURANCE
                                                          63326
                                             11
```

For run time 2500 which model id has idea auc value



Views

Views are stored queries that when invoked produce a result set. A view acts as a virtual table.and Views are used to restrict a data

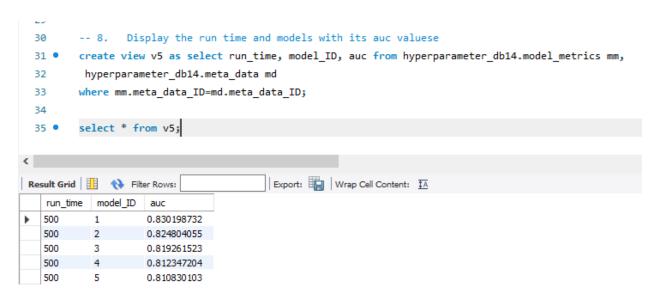
The following illustrates the syntax for creating a new function:

```
CREATE
  [ALGORITHM = {MERGE | TEMPTABLE | UNDEFINED}]
VIEW view_name [(column_list)]
AS
select-statement;
```

Below are the views for the database:

```
3
         -- 1. Display run time, stat time and end time for travel insurance dataset
         create view v1 as select run_time, start_time, end_time from hyperparameter_db14.meta_data;
   4 •
         select * from v1;
   5 •
   6
   7
                                        Export: Wrap Cell Content: IA
run_time start_time
                       end time
            1555918564
                      1555918564
> 500
   1000
           1555920195 1555920195
   1500
            1555968866
                      1555968866
   2000
           1555972904
                      1555972904
   2500
           1555975812
                      1555975812
        -- 2. Display all the models for run time 1500
        create view v2 as select model_ID from hyperparameter_db14.model_metrics mm, hyperparameter_db14.meta_data md
  8 •
  9
        where mm.meta_data_ID=md.meta_data_ID
        and run_time=1500;
 10
 11
        select * from v2;
 12 •
 13
<
                                                                                                             Export: Wrap Cell Content: TA
Result Grid Filter Rows:
   model_ID
1
   2
   3
   4
   5
          -- 3. Create a view having run time and its models where RMSE value is greater than 0.5
  15
          create view v3 as select run_time, model_ID from hyperparameter_db14.model_metrics mm,
  16
          hyperparameter_db14.meta_data md where mm.meta_data_ID=md.meta_data_ID
  17
  18
          and RMSE > 0.05;
  19
         select * from v3;
  20 •
Export: Wrap Cell Content: IA
    run_time | model_ID
  500
   500
   500
   500
   500
```

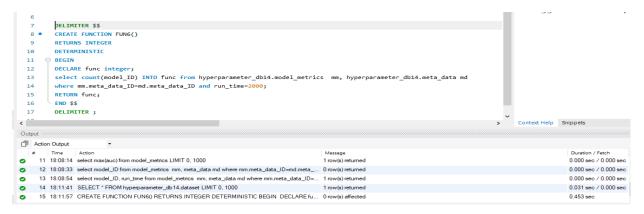
```
22
        -- 5. Display all the metrics values for run time 500
  23 •
        create view v4 as select auc,logloss,mean per class error,rmse,mse from hyperparameter db14.model metrics m ,
        hyperparameter_db14.meta_data_md where m.meta_data_ID=md.meta_data_ID
  24
  25
        and run_time=500;
 26
        select * from v4;
 27 •
  28
<
                                                                                                              Export: Wrap Cell Content: IA
                         mean_per_class_error rmse
0.830198732 0.067567692 0.368385551
                                         0.121168588 0.014681827
  0.824804055 0.065958618 0.364623942
                                      0.119122526 0.014190176
   0.819261523 0.066604943 0.395916695
                                         0.119456081 0.014269755
   0.812347204  0.067957951  0.347793663
                                        0.119143168 0.014195094
  0.118158943 0.013961536
```



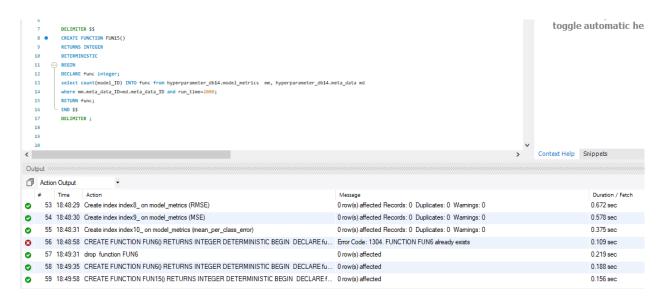
Indexes

Below are the indexes for the database:

1 Create index index1_ on dataset (dataset_ID);

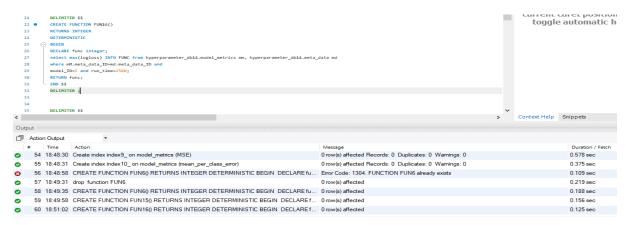


AFTER



2 Create index index2_ on meta_data (meta_data_ID);

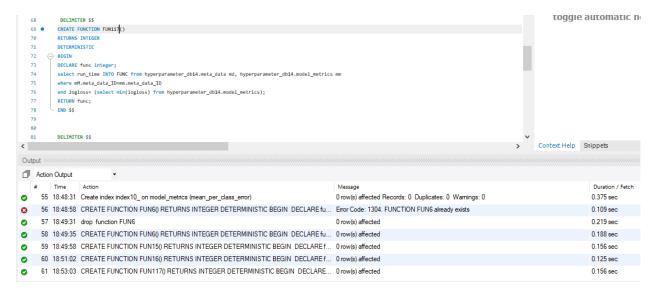




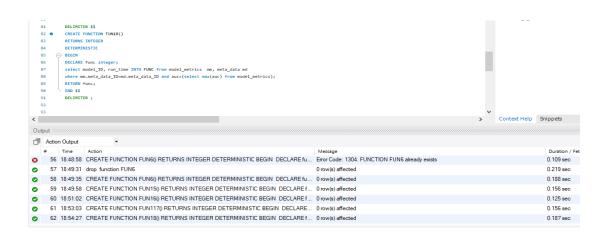
3 Create index index3_ on model_metrics (model_metrics_ID);



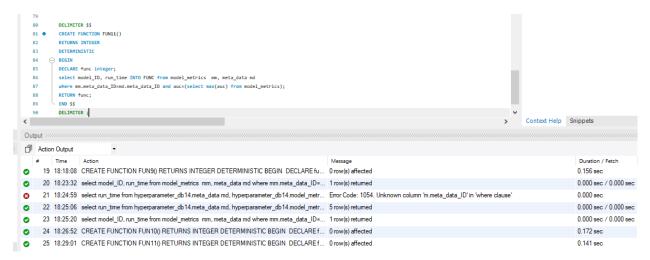
AFTER



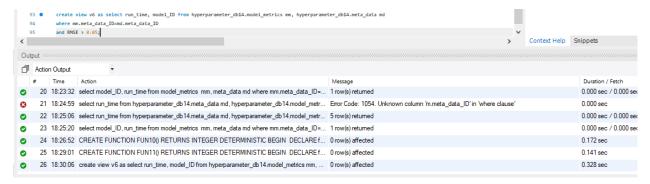
4 Create index index4_ on hyperparameter (hp_ID);



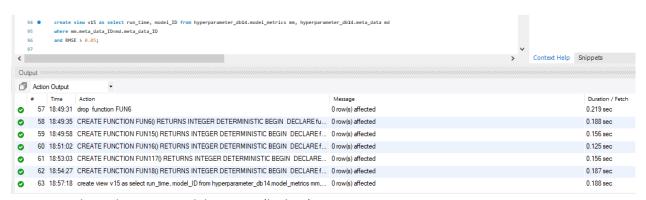
AFTER



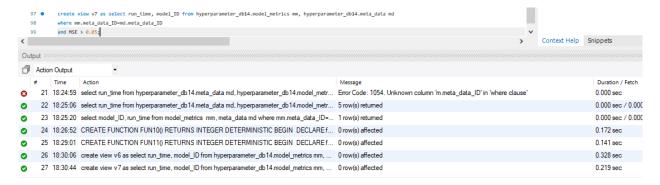
5 Create index index5_ on meta_data(run_time);



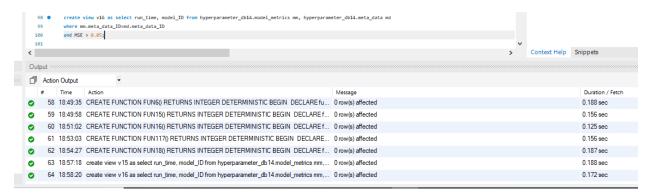
AFTER



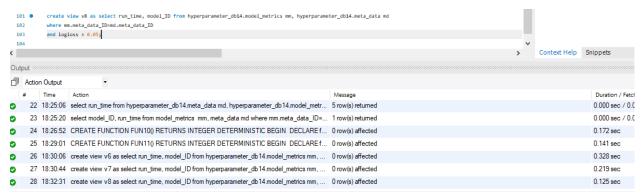
6 Create index index6_ on model_metrics (logloss);

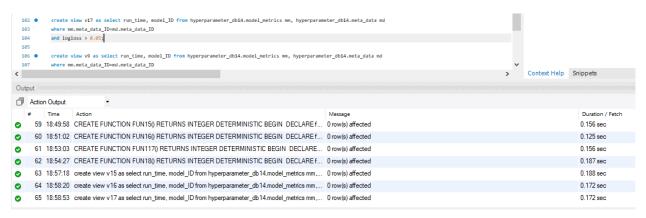


AFTER

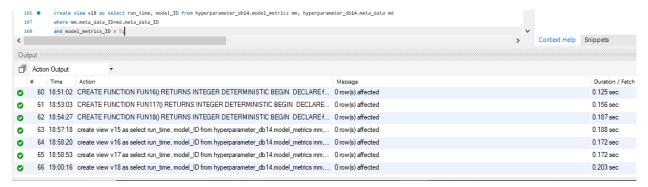


7 Create index index7_ on model_metrics (auc);

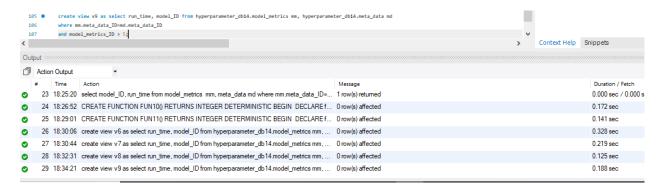




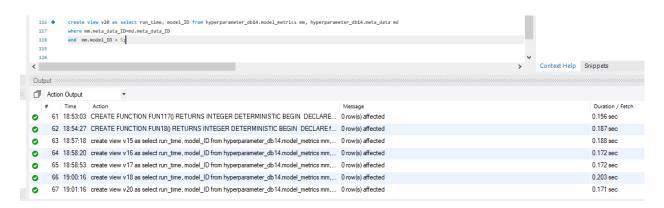
8 Create index index8_ on model_metrics (RMSE);

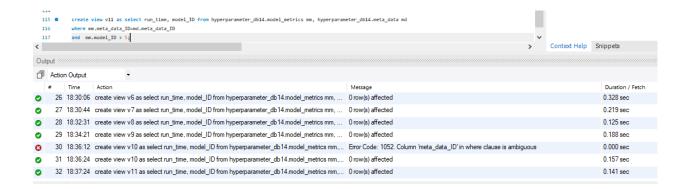


AFTER

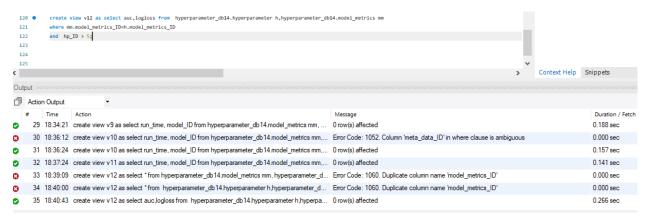


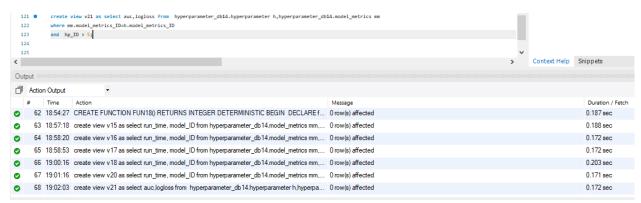
9 Create index index9_ on model_metrics (MSE);

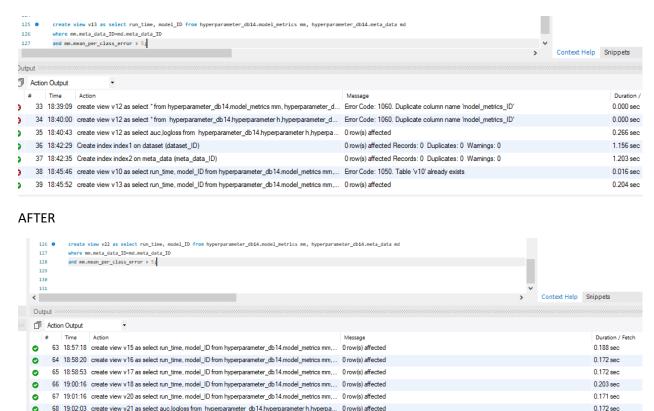




10 Create index index10_ on model_metrics (mean_per_class_error);







Conclusion

Hyperparameter database is built by running several hyperparameter values on Travel Insurance dataset. Data fetched from the dataset is structured into different tables analyzing the relationships between the values.

0.172 sec

Using H20 software, for five run times (500, 1000, 1500, 2000, 2500), respective models and its metrics values are generated. Metrics indicate the absolute fit of the model to actual data. Smaller the metrics values, closer we are to finding the line of best fit. Thus, models generated their own set of hyperparameters. Hence, model architecture is defined for storing the hyperparameter values in the database.

References

https://en.wikipedia.org/wiki/Hyperparameter (machine learning)

69 19:02:38 create view v22 as select run_time, model_ID from hyperparameter_db14.model_metrics mm,... 0 row(s) affected

- https://towardsdatascience.com/top-sources-for-machine-learning-datasetsbb6d0dc3378b
- https://en.wikipedia.org/wiki/List_of_datasets_for_machine-learning_research

- https://medium.com/@alexandraj777/top-5-mistakes-data-scientists-make-with-hyperparameter-optimization-and-how-to-prevent-them-767638b245f8
- https://stats.stackexchange.com/questions/297337/what-are-some-of-the-disavantageof-bayesian-hyper-parameter-optimization
- https://github.com/skunkworksneu/Projects
- http://uksanjay.blogspot.com/2012/06/difference-between-conceptual-logical.html
- https://www.smartdraw.com/entity-relationship-diagram/
- http://www.mysqltutorial.org/create-sql-views-mysql.aspx
- http://www.mysqltutorial.org/mysql-stored-function/
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- https://dev.mysql.com/doc/refman/8.0/en/create-procedure.html

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